

Invariance Principles for Tempered Fractionally Integrated Processes

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Abstract

We discuss invariance principles for autoregressive tempered fractionally integrated moving averages in α -stable ($1 < \alpha \leq 2$) i.i.d. innovations and related tempered linear processes with vanishing tempering parameter $\lambda \sim \lambda_*/N$. We show that the limit of the partial sums process takes a different form in the weakly tempered ($\lambda_* = 0$), strongly tempered ($\lambda_* = \infty$), and moderately tempered ($0 < \lambda_* < \infty$) cases. These results are used to derive the limit distribution of the OLS estimate of AR(1) unit root with weakly, strongly, and moderately tempered moving average errors.

Keywords: invariance principle; tempered linear process; autoregressive fractionally integrated moving average; tempered fractional stable/Brownian motion; tempered fractional unit root distribution;

1 Introduction

The present paper discusses partial sums limits and invariance principles for tempered moving averages

$$X_{d,\lambda}(t) = \sum_{k=0}^{\infty} e^{-\lambda k} b_d(k) \zeta(t-k), \quad t \in \mathbb{Z} \quad (1.1)$$

in i.i.d. innovation process $\{\zeta(t)\}$ with coefficients $b_d(k)$ regularly varying at infinity as k^{d-1} , viz.

$$b_d(k) \sim \frac{c_d}{\Gamma(d)} k^{d-1}, \quad k \rightarrow \infty, \quad c_d \neq 0, \quad d \neq 0 \quad (1.2)$$

where $d \in \mathbb{R}$ is a real number, $d \neq -1, -2, \dots$ and $\lambda > 0$ is tempering parameter. In addition to (1.2) we assume that

$$\sum_{k=0}^{\infty} k^j b_d(k) = 0, \quad 0 \leq j \leq [-d], \quad -\infty < d < 0, \quad (1.3)$$

$$\sum_{k=0}^{\infty} |b_d(k)| < \infty, \quad \sum_{k=0}^{\infty} b_d(k) \neq 0, \quad d = 0 \quad (1.4)$$

An important example of such processes is the two-parametric class ARTFIMA(0, d , λ , 0) of tempered fractionally integrated processes, generalizing the well-known ARFIMA(0, d , 0) class, written as

$$X_{d,\lambda}(t) = (1 - e^{-\lambda}B)^{-d}\zeta(t) = \sum_{k=0}^{\infty} e^{-\lambda k} \omega_{-d}(k) \zeta(t-k), \quad t \in \mathbb{Z} \quad (1.5)$$

with coefficients given by power expansion $(1 - e^{-\lambda}z)^{-d} = \sum_{k=0}^{\infty} e^{-\lambda k} \omega_{-d}(k) z^k$, $|z| < 1$, where $Bx(t) = x(t-1)$ is the backward shift. Due to the presence of the exponential tempering factor $e^{-\lambda k}$ the series in (1.1) and (1.5) absolutely converges a.s. under general assumptions on the innovations, and defines a strictly stationary process. On the other hand, for $\lambda = 0$ the corresponding stationary processes in (1.1) and (1.5) exist under additional conditions on the parameter d . See Granger and Joyeux [12], Hosking [13], Brockwell and Davis [5], Kokoszka and Taqqu [15]. We also note (see e.g. [10], Ch. 3.2) that the (untempered) linear process $X_{d,0}$ of (1.1) with coefficients satisfying (1.2) for $0 < d < 1/2$ is said long memory, while (1.2) and (1.3) for $-1/2 < d < 0$ is termed negative memory and (1.4) short memory, respectively, parameter d usually referred to as memory parameter.

The model in (1.5) appeared in Giraitis et al. [8], which noted that for small $\lambda > 0$, $X_{d,\lambda}$ has a covariance function which resembles the covariance function of a long memory model for arbitrary large number of lags but eventually decays exponentially fast. [8] termed such behavior ‘semi long-memory’ and noted that it may have empirical relevance for modelling of financial returns. Giraitis et al. [9] propose the semi-long memory ARCH(∞) model as a contiguous alternative to (pure) hyperbolic and exponential decay which are often very hard to distinguish between in a finite sample. On the other side, Meerschaert et al. [20] effectively apply ARTFIMA(0, d , λ , 0) in (1.5) for modeling of turbulence in the Great Lakes region.

The present paper obtains limiting behavior of tempered linear processes in (1.1) with small tempering parameter $\lambda = \lambda_N \rightarrow 0$ tending to zero together with the sample size. The important statistic is the partial sums process

$$S_N^{d,\lambda}(t) := \sum_{k=1}^{[Nt]} X_{d,\lambda}(k), \quad t \in [0, 1] \quad (1.6)$$

of $X_{d,\lambda}$ in (1.1) with i.i.d. innovations $\{\zeta(t)\}$ in the domain of attraction of α -stable law, $1 < \alpha \leq 2$. Functional limit theorems for the partial sums process play a crucial role in the R/S analysis, unit root testing, change-point analysis and many other time series inferences. See Lo [17], Phillips [21], Giraitis et al. [9], Lavancier et al. [16] and the references therein.

We prove that the limit behavior of (1.6) essentially depends on how fast $\lambda = \lambda_N$ tends to 0. Assume that there exists the limit

$$\lim_{N \rightarrow \infty} N\lambda_N = \lambda_* \in [0, \infty]. \quad (1.7)$$

Depending on the value of λ_* , the process X_{d,λ_N} will be called *strongly tempered* if $\lambda_* = \infty$, *weakly tempered* if $\lambda_* = 0$, and *moderately tempered* if $0 < \lambda_* < \infty$. While the behavior of S_N^{d,λ_N} in the strongly and weakly tempered cases is typical for short memory and long memory processes, respectively, the moderately tempered decay $\lambda_N \sim \lambda_*/N$, $\lambda_* \in (0, \infty)$ leads to *tempered fractional stable motion of second kind* (TFSM II) Z_{H,α,λ_*}^H , $H = d + 1/\alpha > 0$ defined as a stochastic integral

$$Z_{H,\alpha,\lambda}^H(t) := \int_{\mathbb{R}} h_{H,\alpha,\lambda}(t; y) M_\alpha(y), \quad t \in \mathbb{R} \quad (1.8)$$

with respect to α -stable Lévy process M_α with integrand

$$\begin{aligned} h_{H,\alpha,\lambda}(t; y) &:= (t - y)_+^{H - \frac{1}{\alpha}} e^{-\lambda(t-y)_+} - (-y)_+^{H - \frac{1}{\alpha}} e^{-\lambda(-y)_+} \\ &+ \lambda \int_0^t (s - y)_+^{H - \frac{1}{\alpha}} e^{-\lambda(s-y)_+} ds, \quad y \in \mathbb{R}. \end{aligned} \quad (1.9)$$

TFSM II and its Gaussian counterpart *tempered fractional Brownian motion of second kind* (TFBM II) were recently introduced in Sabzikar and Surgailis [22], the above processes being closely related to the *tempered fractional stable motion* (TFSM) and the *tempered fractional Brownian motion* (TFBM) defined in Meerschaert and Sabzikar [19] and Meerschaert and Sabzikar [18], respectively. As shown in [22], TFSM and TFSM II are different processes, especially striking are their differences as $t \rightarrow \infty$.

As an application of our invariance principles we obtain the limit distribution of the OLS estimator $\hat{\beta}_N$ of the slope parameter in AR(1) model with tempered ARTFIMA(0, d , λ_N , 0) errors and small tempering parameter $\lambda_N \rightarrow 0$ satisfying (1.7), under the null (unit root) hypothesis $\beta = 1$. In the case of (untempered) ARFIMA(0, d , 0) error process with finite variance and standardized i.i.d. innovations, Sowell [24] proved that the distribution of the normalized statistic $N^{1 \wedge (1+2d)}(\hat{\beta}_N - 1)$ tends to the so-called *fractional unit root distribution* written in terms of fractional Brownian motion with parameter $H = d + \frac{1}{2}$. Sowell's [24] result extends the classical unit root distribution for weakly dependent errors in Phillips [21] to fractionally integrated error process, yielding drastically different limits for $0 < d < 1/2$, $d = 0$ and $-1/2 < d < 0$.

It turns out that in the case of ARTFIMA(0, d , λ_N , 0) error process with $\lambda_N \sim \lambda_*/N$, the limit distribution of $\hat{\beta}_N$ depends on $\lambda_* \in [0, \infty]$ and d . Roughly speaking (see Theorem 5.2 for precise formulation), in the moderately tempered case $0 < \lambda_* < \infty$ the limit distribution of $\hat{\beta}_N$ writes similarly to Sowell [24] with FBM B_H replaced by TFBM II B_{H,λ_*}^H and the convergence holds for all $-1/2 < d < \infty$ in contrast to [24] which is limited to $|d| < 1/2$. Under strong tempering $\lambda_N/N \rightarrow \lambda_* = \infty$, the limit distribution of $\hat{\beta}_N$ is written in terms of standard Brownian motion but takes a different form in the cases $d > 0$, $d = 0$ and $d < 0$, $d \neq \mathbb{N}_-$; moreover, except for the i.i.d. case $d = 0$, this limit is different from Sowell's limit in [24] and also from the unit root distribution in Dickey and Fuller [6] and Phillips [21].

The paper is organized as follows. Section 2 introduces ARTFIMA(p, d, λ, q) class and provides basic properties of these processes. In Section 3 we define TFSM II/TFBM II. Section 4 contains the main results of the paper (invariance principles). Section 5 discusses the application to unit root testing. The proofs of the main results are relegated to Section 6.

In what follows, C denotes generic constants which may be different at different locations. We write \xrightarrow{d} , $\stackrel{d}{=}$, $\xrightarrow{\text{fdd}}$, $\stackrel{\text{fdd}}{=}$ for the weak convergence and equality of distributions and finite-dimensional distributions. $\mathbb{N}_\pm := \{\pm 1, \pm 2, \dots\}$, $\mathbb{R}_+ := (0, \infty)$, $(x)_\pm := \max(\pm x, 0)$, $x \in \mathbb{R}$, $\int := \int_{\mathbb{R}}$. $L^p(\mathbb{R})$ ($p \geq 1$) denotes the Banach space of measurable functions $f : \mathbb{R} \rightarrow \mathbb{R}$ with finite norm $\|f\|_p = (\int |f(x)|^p dx)^{1/p}$.

2 Tempered fractionally integrated process

In this section, we define ARTFIMA(p, d, λ, q) process and discuss its basic properties. Let $\Phi(z) = 1 - \sum_{i=1}^p \phi_i z^i$ and $\Theta(z) = 1 + \sum_{i=1}^q \theta_i z^i$ be polynomials with real coefficients of degree $p, q \geq 0$, such that $\Phi(z)$ does not vanish on $\{z \in \mathbb{C}, |z| \leq 1\}$ and $\Phi(z)$ and $\Theta(z)$ have no common zeros. Let $d \in \mathbb{R} \setminus \mathbb{N}_+$. Consider Taylor's expansion

$$\frac{\Theta(z)}{\Phi(z)}(1-z)^d = \sum_{k=0}^{\infty} a_d(k) z^k, \quad |z| < 1. \quad (2.1)$$

Note that

$$a_d(k) = \sum_{s=0}^k \omega_d(k) \psi(k-s), \quad k \geq 0, \quad (2.2)$$

where $\omega_d(k) = \frac{\Gamma(k-d)}{\Gamma(k+1)\Gamma(d)}$, and $\psi(j)$ are the coefficients of the power series $\sum_{j=0}^{\infty} \psi(j) z^j = \Theta(z)/\Phi(z)$, $|z| \leq 1$. We use the fact (see Kokoszka and Taqqu ([15], Lemma 3.1) that for any $d \in \mathbb{R} \setminus \mathbb{N}_-$

$$\omega_{-d}(k) = \frac{\Gamma(k+d)}{\Gamma(k+1)\Gamma(d)} = \Gamma(d)^{-1} k^{d-1} (1 + O(1/k)), \quad k \rightarrow \infty. \quad (2.3)$$

Proposition 2.1 *Let $\Theta(1) \neq 0$. For any $d \in \mathbb{R} \setminus \mathbb{N}_-$, the coefficients $a_{-d}(k)$, $k \geq 0$ satisfy conditions (1.2)-(1.4). In particular, for any $d \in \mathbb{R} \setminus \{0, -1, -2, \dots\}$*

$$a_{-d}(k) \sim \frac{\Theta(1)}{\Phi(1)\Gamma(d)} k^{d-1} \left(1 + O\left(\frac{1}{k}\right)\right), \quad k \rightarrow \infty. \quad (2.4)$$

Proof. Let us prove (2.4). By definition, $a_{-d}(k) = \sum_{j=0}^k \psi(j) \omega_{-d}(k-j)$, $k \geq 0$, see (2.2). It is well-known that $|\psi(j)| \leq C e^{-cj}$ for some constants $C, c > 0$, see ([15], proof of Lemma 3.2). Note $\Theta(1)/\Phi(1) = \sum_{j=0}^{\infty} \psi(j) \neq 0$. We have

$$\begin{aligned} & |a_{-d}(k) - (\Theta(1)/\Phi(1)) \omega_{-d}(k)| \\ &= \left| \sum_{j=0}^k \psi(j) (\omega_{-d}(k-j) - \omega_{-d}(k)) - \omega_{-d}(k) \sum_{j>k} \psi(j) \right| \leq \sum_{i=1}^3 \ell_{k,i}, \end{aligned}$$

where

$$\begin{aligned}\ell_{k,1} &:= \sum_{0 \leq j \leq k^{1/4}} |\psi(j)| |\omega_{-d}(k-j) - \omega_{-d}(k)|, \\ \ell_{k,2} &:= \sum_{k^{1/4} < j \leq k} |\psi(j)| (|\omega_{-d}(k-j)| + |\omega_{-d}(k)|),\end{aligned}$$

and $\ell_{k,3} := |\omega_{-d}(k)| \sum_{j > k} |\psi(j)| \leq Ck^{d-1} \sum_{j > k} e^{-cj} = o(k^{d-2})$ since $\psi(j)$ decay exponentially. Similarly, $\ell_{k,2} \leq Ce^{-ck^{1/4}} k \max_{1 \leq j \leq k} |\omega_{-d}(j)| = o(k^{d-2})$. Using (2.3) we obtain $|\ell_{k,1}| \leq C(\ell'_{k,1} + \ell''_{k,2})$, where $\ell''_{k,2} := k^{d-2} \sum_{j \geq 0} |\psi(j)| = O(k^{d-2})$ and

$$\begin{aligned}\ell'_{k,1} &:= \sum_{0 \leq j \leq k^{1/4}} |\psi(j)| (k^{d-1} - (k-j)^{d-1}) = k^{d-1} \sum_{0 \leq j \leq k^{1/2}} |\psi(j)| \left(1 - \left(1 - \frac{j}{k}\right)^{d-1}\right) \\ &\leq Ck^{d-2} \sum_{j \geq 0} j |\psi(j)| = O(k^{d-2})\end{aligned}$$

proving $|a_{-d}(k) - (\Theta(1)/\Phi(1))\omega_{-d}(k)| = O(k^{d-2})$ and hence (2.4) in view of (2.3). Thus, $a_{-d}(k)$, $d \neq 0$ satisfy (1.2). Condition (1.4) is obvious from $a_0(k) = \psi(k)$ and properties of $\psi(k)$ stated above.

It remains to prove (1.3). Let $j < -d < j+1$ for some $j = 0, 1, \dots$. Then since $\Psi(z) := \Theta(z)/\Phi(z)$ is analytic on $\{z \in \mathbb{C}, |z| \leq 1+\delta\}$ ($\exists \delta > 0$) and $|a_{-d}(k)| \leq Ck^{d-1}$, see (2.4), the function $\sum_{k=0}^{\infty} a_{-d}(k)z^k = \Psi(z)(1-z)^{-d}$ is j times differentiable on $\{|z| \leq 1\}$ and $\frac{\partial^i}{\partial z^i} \Psi(z)(1-z)^{-d} \Big|_{z=1} = \sum_{r=0}^i \binom{i}{r} \frac{\partial^r \Psi(z)}{\partial z^r} \Big|_{z=1} \frac{\partial^{i-r}(1-z)^{-d}}{\partial z^{i-r}} \Big|_{z=1} = 0, 0 \leq i \leq j$. Hence, $0 = \sum_{k=i}^{\infty} k(k-1) \cdots (k-i+1) a_{-d}(k) = \sum_{k=0}^{\infty} k(k-1) \cdots (k-i+1) a_{-d}(k)$ for any $0 \leq i \leq j$, proving (1.3) and the proposition, too. \square

Definition 2.2 Let the autoregressive polynomials $\Theta(z), \Phi(z)$ of degree p, q satisfy the above conditions, and $d \in \mathbb{R} \setminus \mathbb{N}_-, \lambda \geq 0$. Moreover, let $\zeta = \{\zeta(t), t \in \mathbb{Z}\}$ be a stationary process with $\mathbb{E}|\zeta(0)| < \infty$. By $\text{ARTFIMA}(p, d, \lambda, q)$ process with innovation process ζ we mean a stationary moving-average process $X_{p,d,\lambda,q} = \{X_{p,d,\lambda,q}(t), t \in \mathbb{Z}\}$ defined by

$$X_{p,d,\lambda,q}(t) = \sum_{k=0}^{\infty} e^{-\lambda k} a_{-d}(k) \zeta(t-k), \quad t \in \mathbb{Z} \quad (2.5)$$

where the series converges in L_1 .

Remark 2.3 (i) For $\lambda = 0$ and $|d| < 1/2$ and i.i.d. innovations with zero mean and unit variance, $\text{ARTFIMA}(p, d, 0, q)$ process $X_{p,d,0,q}$ coincides with $\text{ARFIMA}(p, d, q)$ process, see e.g. Brockwell and Davis [5]. Particularly, $X_{d,0} = X_{0,d,0,0}$ in (2.5) is a stationary solution of the $\text{AR}(\infty)$ equation

$$(1-B)^d X_{d,0}(t) = \sum_{j=0}^{\infty} \omega_d(j) X_{d,0}(t-j) = \zeta(t) \quad (2.6)$$

and the series in (2.5) and (2.6) converge in L_2 , meaning that $X_{d,0}$ is invertible.

(ii) For $\lambda = 0$ and zero mean i.i.d. α -stable innovations, $1 < \alpha < 2$, the definition of $X_{p,d,0,q}$ in Definition 2.2 agrees with the definition of ARFIMA(p, d, q) process in [15], who showed that the series in (2.5) converges a.s. and in L_1 for $d < 1 - \frac{1}{\alpha}$.

Proposition 2.4 *Let $\Theta(z), \Phi(z)$ satisfy the above conditions and $\lambda > 0$. Then the series in (2.5) converges in L_1 for any $d \in \mathbb{R} \setminus \mathbb{N}_-$ hence ARTFIMA(p, d, λ, q) process $X_{p,d,\lambda,q}$ in Definition 2.2 is well-defined for arbitrary (stationary) innovation process ζ with finite mean. Moreover, if $|\Theta(z)| > 0, |z| \leq 1$ then $X_{p,d,\lambda,q}$ is invertible:*

$$\zeta(t) = \sum_{k=0}^{\infty} e^{-\lambda k} \tilde{a}_d(k) X_{p,d,\lambda,q}(t-k), \quad (2.7)$$

where

$$\sum_{k=0}^{\infty} \tilde{a}_d(k) z^k = \frac{\Phi(z)}{\Theta(z)} (1-z)^d, \quad |z| < 1 \quad (2.8)$$

and the series in (2.7) converges in L_1 .

Proof. The convergence in L_1 of the series in (2.5) follows from (2.4). To show invertibility of these series, note that by Proposition 2.1 the coefficients in (2.8) satisfy the bound $|\tilde{a}_d(k)| \leq Ck^{-d-1}, k \geq 1$ for $d \in \mathbb{R} \setminus \mathbb{N}_+$, and an exponential bound $|\tilde{a}_d(k)| \leq Ce^{-ck}, k \geq 1, c > 0$ for $d \in \mathbb{N}_+$. Hence, $e^{-\lambda k} |\tilde{a}_d(k)| \leq Ce^{-\lambda k} k^{-d-1}$, for any $d \in \mathbb{R}$ implying the convergence in L_1 of the series in (2.7). Finally, equality in (2.7) follows from identity $1 = (\frac{\Phi(z)}{\Theta(z)}(1-z)^d)(\frac{\Theta(z)}{\Phi(z)}(1-z)^{-d}), |z| < 1$. \square

Proposition 2.5 describes some second order properties of ARTFIMA(p, d, λ, q) with standardized innovations.

Proposition 2.5 *Let $X_{p,d,\lambda,q}$ be ARTFIMA(p, d, λ, q) process in (2.5), $d \in \mathbb{R} \setminus \mathbb{N}_-, \lambda > 0$, with standardized i.i.d. innovations $\{\zeta(t), t \in \mathbb{Z}\}, \mathbb{E}\zeta(0) = 0, \mathbb{E}\zeta^2(0) = 1$. Then $\mathbb{E}X_{p,d,\lambda,q}(t) = 0, \mathbb{E}X_{p,d,\lambda,q}^2(t) = \sum_{k=0}^{\infty} e^{-2\lambda k} a_{-d}^2(k) < \infty$ and*

(i) *The spectral density of $X_{p,d,\lambda,q}$ is given by*

$$h(x) = \frac{1}{2\pi} \left| \frac{\Theta(e^{-ix})}{\Phi(e^{-ix})} \right|^2 (1 - 2e^{-\lambda} \cos x + e^{-2\lambda})^{-d}, \quad -\pi \leq x \leq \pi.$$

(ii) *The covariance function of $X_{0,d,\lambda,0}$ is given by*

$$\gamma_{d,\lambda}(k) = \mathbb{E}X_{0,d,\lambda,0}(0)X_{0,d,\lambda,0}(k) = \frac{e^{-\lambda k} \Gamma(d+k)}{\Gamma(d)\Gamma(k+1)} {}_2F_1(d, k+d; k+1; e^{-2\lambda}), \quad (2.9)$$

where ${}_2F_1(a, b; c; z)$ is the Gauss hypergeometric function (see e.g. [11]). Moreover,

$$\sum_{k \in \mathbb{Z}} |\gamma_{d,\lambda}(k)| < \infty, \quad \sum_{k \in \mathbb{Z}} \gamma_{d,\lambda}(k) = (1 - e^{-\lambda})^{-2d} \quad (2.10)$$

and

$$\gamma_{d,\lambda}(k) \sim Ak^{d-1}e^{-\lambda k}, \quad k \rightarrow \infty, \quad \text{where } A = (1 - e^{-2\lambda})^{-d}\Gamma(d)^{-1}. \quad (2.11)$$

Proof. (i) From the transfer function $(\Theta(e^{-ix})/\Phi(e^{-ix}))(1 - e^{-ix-\lambda})^{-d}$ of the filter in (2.5) we have that $h(x) = \frac{1}{2\pi} |\Theta(e^{-ix})/\Phi(e^{-ix})|^2 |1 - e^{-ix-\lambda}|^{-2d}$, where $|1 - e^{-ix-\lambda}|^2 = 1 - 2e^{-\lambda} \cos(x) + e^{-2\lambda}$, proving part (i). (ii) (2.9) follows from $\gamma_{d,\lambda}(k) = \int_{-\pi}^{\pi} \cos(kx) h(x) dx$ and ([11], Eq. 9.112). The first relation in (2.10) follows from $\sum_{j \in \mathbb{Z}} |e^{-\lambda j} \omega_d(j)| < \infty$, see (2.3), and the second one from $\sum_{k \in \mathbb{Z}} \gamma_{d,\lambda}(k) = 2\pi h(0) = (1 - e^{-\lambda})^{-2d}$. Finally, (2.11) is proved in ([8], (4.15)). \square

3 Tempered fractional Brownian and stable motions of second kind

This section contains the definition of TFBM II/TFSM II and some of its properties from Sabzikar and Surgailis [22]. The reader is referred to the aforementioned paper for further properties of these processes including relation to tempered fractional calculus, relation between TFBM II/TFSM II and TFBM/TFSM, dependence properties of the increment process (tempered fractional Brownian/stable noise), local and global asymptotic self-similarity.

For $1 < \alpha \leq 2$, let $M_\alpha = \{M_\alpha(t), t \in \mathbb{R}\}$ be an α -stable Lévy process with stationary independent increments and characteristic function

$$\mathbb{E} e^{i\theta M_\alpha(t)} = e^{-\sigma^\alpha |\theta|^\alpha |t| \left(1 - \beta \tan(\pi\alpha/2) \text{sign}(\theta)\right)}, \quad \theta \in \mathbb{R}, \quad (3.1)$$

where $\sigma > 0$ and $\beta \in [-1, 1]$ are the scale and skewness parameters, respectively. For $\alpha = 2$, $M_2(t) = \sqrt{2}\sigma B(t)$, where B is a standard Brownian motion with variance $\mathbb{E} B^2(t) = t$. Stochastic integral $I_\alpha(f) \equiv \int f(x) M_\alpha(dx)$ is defined for any $f \in L^\alpha(\mathbb{R})$ as α -stable random variable with characteristic function

$$\mathbb{E} e^{i\theta I_\alpha(f)} = \exp\{-\sigma^\alpha |\theta|^\alpha \int |f(x)|^\alpha (1 - \beta \tan(\pi\alpha/2) \text{sign}(\theta f(x))) dx\}, \quad \theta \in \mathbb{R}. \quad (3.2)$$

see e.g. [23, Chapter 3].

For $t \in \mathbb{R}, H > 0, 1 < \alpha \leq 2, \lambda \geq 0$ consider the function $y \mapsto h_{H,\alpha,\lambda}(t; y) : \mathbb{R} \rightarrow \mathbb{R}$ given in (1.9). Note $h_{H,\alpha,\lambda}(t; \cdot) \in L^\alpha(\mathbb{R})$ for any $t \in \mathbb{R}, \lambda > 0, 1 < \alpha \leq 2, H > 0$ and also for $\lambda = 0, 1 < \alpha \leq 2, H \in (0, 1)$. We will use the following integral representation of (1.9) (see [22]). For $H > \frac{1}{\alpha}$:

$$h_{H,\alpha,\lambda}(t; y) = (H - \frac{1}{\alpha}) \int_0^t (s - y)_+^{H - \frac{1}{\alpha} - 1} e^{-\lambda(s-y)_+} ds. \quad (3.3)$$

For $0 < H < \frac{1}{\alpha}$:

$$h_{H,\alpha,\lambda}(t; y) = (H - \frac{1}{\alpha}) \begin{cases} \int_0^t (s - y)_+^{H - \frac{1}{\alpha} - 1} e^{-\lambda(s-y)_+} ds, & y < 0, \\ -\int_t^\infty (s - y)_+^{H - \frac{1}{\alpha} - 1} e^{-\lambda(s-y)_+} ds + \lambda^{\frac{1}{\alpha} - H} \Gamma(H - \frac{1}{\alpha}), & y \geq 0. \end{cases} \quad (3.4)$$

Definition 3.1 Let M_α be α -stable Lévy process in (3.1), $1 < \alpha \leq 2$ and $H > 0$, $\lambda > 0$. Define, for $t \in \mathbb{R}$, the stochastic integral

$$Z_{H,\alpha,\lambda}^{\text{II}}(t) := \int h_{H,\alpha,\lambda}(t; y) M_\alpha(y). \quad (3.5)$$

The process $Z_{H,\alpha,\lambda}^{\text{II}} = \{Z_{H,\alpha,\lambda}^{\text{II}}(t), t \in \mathbb{R}\}$ will be called *tempered fractional stable motion of second kind* (TFSM II). A particular case of (3.5) corresponding to $\alpha = 2$

$$B_{H,\lambda}^{\text{II}}(t) := \frac{1}{\Gamma(H + \frac{1}{2})} \int h_{H,2,\lambda}(t; y) B(y) \quad (3.6)$$

will be called *tempered fractional Brownian motion of second kind* (TFBM II).

The next proposition states some basic properties of TFSM II $Z_{H,\alpha,\lambda}^{\text{II}}$.

Proposition 3.2 ([22])

- (i) $Z_{H,\alpha,\lambda}^{\text{II}}$ in (3.5) is well-defined for any $t \in \mathbb{R}$ and $1 < \alpha \leq 2, H > 0, \lambda > 0$, as a stochastic integral in (3.2).
- (ii) $Z_{H,\alpha,\lambda}^{\text{II}}$ in (3.5) has stationary increments and α -stable finite-dimensional distributions. Moreover, it satisfies the following scaling property: $\{Z_{H,\alpha,\lambda}^{\text{II}}(bt), t \in \mathbb{R}\} \stackrel{\text{fdd}}{=} \{b^H Z_{H,\alpha,b\lambda}^{\text{II}}(t), t \in \mathbb{R}\}$, $\forall b > 0$.
- (iii) $Z_{H,\alpha,\lambda}^{\text{II}}$ in (3.5) has a.s. continuous paths if either $\alpha = 2, H > 0$, or $1 < \alpha < 2, H > 1/\alpha$ hold.

4 Invariance principles

In this section, we discuss invariance principles and the convergence of (normalized) partial sums process

$$S_N^{d,\lambda}(t) := \sum_{k=1}^{[Nt]} X_{d,\lambda}(k), \quad t \in [0, 1] \quad (4.1)$$

of tempered linear process $X_{d,\lambda}$ in (1.1) with i.i.d. innovations belonging to the domain of attraction of α -stable law, $1 < \alpha \leq 2$ (see below). We shall assume that the tempering parameter $\lambda \equiv \lambda_N \rightarrow 0$ as $N \rightarrow \infty$ and following limit exists:

$$\lim_{N \rightarrow \infty} N\lambda_N = \lambda_* \in [0, \infty]. \quad (4.2)$$

Recall that X_{d,λ_N} is said *strongly tempered* if $\lambda_* = \infty$, *weakly tempered* if $\lambda_* = 0$, and *moderately tempered* if $0 < \lambda_* < \infty$. We show that the limits of the partial sums process in (4.1) exist under condition (4.2) and depend on λ_*, d, α ; moreover, in all cases these limits belong to the class of TFSM II processes defined in Section 3.

Definition 4.1 Write $\zeta \in D(\alpha)$, $1 < \alpha \leq 2$ if

- (i) $\alpha = 2$ and $\mathbb{E}\zeta = 0$, $\sigma^2 := \mathbb{E}\zeta^2 < \infty$, or
- (ii) $1 < \alpha < 2$ and there exist some constants $c_1, c_2 \geq 0, c_1 + c_2 > 0$ such that $\lim_{x \rightarrow \infty} x^\alpha \mathbb{P}(\zeta > x) = c_1$ and $\lim_{x \rightarrow -\infty} |x|^\alpha \mathbb{P}(\zeta \leq x) = c_2$; moreover, $\mathbb{E}\zeta = 0$.

Condition $\zeta \in D(\alpha)$ implies that r.v. ζ belongs to the domain of normal attraction of an α -stable law. In other words, if $\zeta(i), i \in \mathbb{Z}$ are i.i.d. copies of ζ then

$$N^{-1/\alpha} \sum_{i=1}^{[Nt]} \zeta(i) \xrightarrow{\text{fdd}} M_\alpha(t), \quad N \rightarrow \infty, \quad (4.3)$$

where M_α is an α -stable Lévy process in (3.1) with σ, β determined by c_1, c_2 , see ([7], pp. 574-581). We shall use the following criterion for convergence of weighted sums in i.i.d. r.v.s. See ([10], Prop. 14.3.2), [2], [14].

Proposition 4.2 Let $1 < \alpha \leq 2$ and $Q(g_N) = \sum_{t \in \mathbb{Z}} g_N(t) \zeta(t)$ be a linear form in i.i.d. r.v.s $\zeta(t) \in D(\alpha)$ with real coefficients $g_N(t), t \in \mathbb{Z}$. Assume that there exists $p \in [1, \alpha)$ if $\alpha < 2$, $p = 2$ if $\alpha = 2$ and a function $g \in L^p(\mathbb{R})$ such that the functions

$$\tilde{g}_N(x) := N^{1/\alpha} g_N([xN]), \quad x \in \mathbb{R} \quad (4.4)$$

satisfy

$$\|\tilde{g}_N - g\|_p \rightarrow 0 \quad (N \rightarrow \infty). \quad (4.5)$$

Then $Q(g_N) \xrightarrow{d} \int g(x) M_\alpha(x)$, where M_α is as in (4.3) and (3.1).

Write $\xrightarrow{D[0,1]}$ for weak convergence of random processes in the Skorohod space $D[0,1]$ equipped with J_1 -topology, see [4].

In Theorem 4.3 below, X_{d,λ_N} is a tempered linear process of (1.1) with i.i.d. innovations $\zeta(t) \in D(\alpha), 1 < \alpha \leq 2$, coefficients $b_d(k), k \geq 0, d \in \mathbb{R} \setminus \mathbb{N}_-$ satisfying (1.2)-(1.4), and tempering parameter $0 < \lambda_N \rightarrow 0 (N \rightarrow \infty)$ satisfying (4.2). W.l.g., we shall assume that the asymptotic constant c_d in (1.2)-(1.4) equals 1: $c_d = 1 \forall d \in \mathbb{R} \setminus \mathbb{N}_-$.

Theorem 4.3 (i) (Strongly tempered process.) Let $\lambda_* = \infty$ and $d \in \mathbb{R} \setminus \mathbb{N}_-$. Then

$$N^{-\frac{1}{\alpha}} \lambda_N^d S_N^{d,\lambda_N}(t) \xrightarrow{\text{fdd}} M_\alpha(t), \quad (4.6)$$

where M_α is α -stable Lévy process in (4.3). Moreover, if $\alpha = 2$ and $\mathbb{E}|\zeta(0)|^p < \infty$ for some $p > 2$ then

$$N^{-\frac{1}{2}} \lambda_N^d S_N^{d,\lambda_N}(t) \xrightarrow{D[0,1]} \sigma B(t), \quad (4.7)$$

where B is a standard Brownian motion and $\sigma > 0$ some constant.

(ii) (Weakly tempered process.) Let $\lambda_* = 0$ and $H = d + \frac{1}{\alpha} \in (0, 1)$. Then

$$N^{-H} S_N^{d, \lambda_N}(t) \xrightarrow{\text{fdd}} \Gamma(d+1)^{-1} Z_{H, \alpha, 0}(t), \quad (4.8)$$

where $Z_{H, \alpha, 0}$ is a linear fractional α -stable motion, see Definition 3.1. Particularly, for $\alpha = 2$, $Z_{H, 2, 0}$ is a multiple of FBM.

Moreover, if either $1 < \alpha \leq 2, 1/\alpha < H < 1$, or $\alpha = 2, 0 < H < 1/2$ and $\mathbb{E}|\zeta(0)|^p < \infty$ ($\exists p > 1/H$) hold, then $\xrightarrow{\text{fdd}}$ in (4.8) can be replaced by $\xrightarrow{D[0,1]}$.

(iii) (Moderately tempered process.) Let $\lambda_* \in (0, \infty)$ and $H = d + \frac{1}{\alpha} > 0$. Then

$$N^{-H} S_N^{d, \lambda_N}(t) \xrightarrow{\text{fdd}} \Gamma(d+1)^{-1} Z_{H, \alpha, \lambda_*}^{\text{II}}(t), \quad (4.9)$$

where $Z_{H, \alpha, \lambda_*}^{\text{II}}$ is a TFMS II as defined in Definition 3.1.

Moreover, if either $1 < \alpha \leq 2, 1/\alpha < H$, or $\alpha = 2, 0 < H < 1/2$ and $\mathbb{E}|\zeta(0)|^p < \infty$ ($\exists p > 1/H$) hold, then $\xrightarrow{\text{fdd}}$ in (4.9) can be replaced by $\xrightarrow{D[0,1]}$.

Remark 4.4 Note that for $\lambda_N = \lambda_*/N$ the normalization in (4.6) becomes $N^{-(\frac{1}{\alpha}+d)}\lambda_*^d$ where the exponent $\frac{1}{\alpha} + d = H$ is the same as in (4.8) and (4.9).

Remark 4.5 The functional convergence in (4.6), case $1 < \alpha < 2$ (the case of discontinuous limit process) is open and apparently does not hold in the usual J_1 -topology, see [3]. In the case of (4.8) and (4.9) and $1 < \alpha < 2, 0 < H < \frac{1}{\alpha}$, functional convergence cannot hold in principle since the limit processes do not belong to $D[0, 1]$.

5 Tempered fractional unit root distribution

A fundamental problem of time series is testing for the unit root $\beta = 1$ in the AR(1) model

$$Y(t) = \beta Y(t-1) + X(t), \quad t = 1, 2, \dots, N, \quad Y(0) = 0 \quad (5.1)$$

with stationary error process $X = \{X(t), t \in \mathbb{Z}\}$. The classical approach to the unit root testing is based on the limit distribution of the OLS estimator $\hat{\beta}_N$

$$\hat{\beta}_N = \frac{\sum_{t=1}^N Y(t)Y(t-1)}{\sum_{t=1}^N Y^2(t-1)}. \quad (5.2)$$

The limit theory for $\hat{\beta}_N$ in the case of weakly dependent errors X was developed in Phillips [21]. We note that [21] makes an extensive use of invariance principle for the error process. Sowell [24] obtained the limit distribution of $\hat{\beta}_N$ in the case of strongly dependent ARFIMA(0, d , 0) error

process with finite variance and standardized i.i.d. innovations. [24] proved that the distribution of the normalized statistic $N^{1 \wedge (1+2d)}(\hat{\beta}_N - 1)$ tends to that of the ratio

$$\frac{1}{2 \int_0^1 B_H^2(s) \mathfrak{s}} \begin{cases} B_H^2(1), & 0 < d < 1/2, \\ B^2(1) - 1, & d = 0, \\ -H\Gamma(H + \frac{1}{2})/\Gamma(\frac{3}{2} - H), & -1/2 < d < 0, \end{cases} \quad (5.3)$$

where $H = d + \frac{1}{2}$ and B_H is a FBM with parameter $H \in (0, 1)$, $B = B_{1/2}$ being a standard Brownian motion.

In this section we extend Sowell's [24] result to ARTFIMA(0, $d, \lambda_N, 0$) error process with small tempering parameter $\lambda_N \sim \lambda_*/N \rightarrow 0$ as in (4.2). Although Theorem 5.2 can be generalized to more general tempered processes with finite variance as in Theorem 4.3, our choice of ARTFIMA(0, $d, \lambda_N, 0$) as the error process is motivated by better comparison to [24]. As noted in Section 1, the degree of tempering has a strong effect on the limit distribution of $\hat{\beta}_N$ and leads to a new two-parameter family of tempered fractional unit root distributions. Following [24], we decompose

$$\hat{\beta}_N - 1 = \hat{A}_N - \hat{B}_N, \quad (5.4)$$

where

$$\hat{A}_N := \frac{Y^2(N)}{2 \sum_{t=1}^N Y^2(t-1)}, \quad \hat{B}_N := \frac{\sum_{t=1}^N X^2(t)}{2 \sum_{t=1}^N Y^2(t-1)}. \quad (5.5)$$

Under the unit root hypothesis $\beta = 1$ we have $Y(t) = \sum_{i=1}^t X(i) = S_N(t/N)$, where $S_N(x) := \sum_{t=1}^{[Nx]} X(t)$, $x \in [0, 1]$ is the partial sums process. Particularly, the statistics in (5.5) can be rewritten as

$$\hat{A}_N = \frac{S_N^2(1)}{2N \int_0^1 S_N^2(s) \mathfrak{s}}, \quad \hat{B}_N = \frac{\sum_{t=1}^N X^2(t)}{2N \int_0^1 S_N^2(s) \mathfrak{s}}. \quad (5.6)$$

For ARTFIMA error process X_{d, λ_N} , the behavior of $S_N^2(1)$ and $\int_0^1 S_N^2(s) \mathfrak{s}$ can be derived from Theorem 4.3. The behavior of $\sum_{t=1}^N X^2(t)$ is established in the following proposition.

Proposition 5.1 *Let X_{d, λ_N} be an ARTFIMA(0, $d, \lambda_N, 0$) process in (2.5) with i.i.d. innovations $\{\zeta(t)\}$, $\mathbb{E}\zeta(0) = 0$, $\mathbb{E}\zeta^2(0) = 1$, fractional parameter $d \in \mathbb{R} \setminus \mathbb{N}_-$ and tempering parameter $\lambda_N \rightarrow 0$. Moreover, let $\mathbb{E}|\zeta(0)|^p < \infty$ ($\exists p > 2$). Then*

$$\frac{1}{N} \sum_{t=1}^N X_{d, \lambda_N}^2(t) \xrightarrow{p} \frac{\Gamma(1-2d)}{\Gamma^2(1-d)}, \quad d < 1/2, \quad (5.7)$$

$$\frac{\lambda_N^{2d-1}}{N} \sum_{t=1}^N X_{d, \lambda_N}^2(t) \xrightarrow{p} \frac{\Gamma(d-1/2)}{2\sqrt{\pi}\Gamma(d)}, \quad d > 1/2, \quad (5.8)$$

$$\frac{1}{N|\log \lambda_N|} \sum_{t=1}^N X_{d, \lambda_N}^2(t) \xrightarrow{p} \frac{1}{\pi}, \quad d = 1/2. \quad (5.9)$$

The main result of this section is the following theorem.

Theorem 5.2 *Consider the AR(1) model in (5.1) with $\beta = 1$ and ARTFIMA(0, $d, \lambda_N, 0$) error process $X = \{X_{d, \lambda_N}(t)\}$ in (2.5) with i.i.d. innovations $\{\zeta(t), t \in \mathbb{Z}\}$, $\mathbb{E}\zeta(0) = 0$, $\mathbb{E}\zeta^2(0) = 1$, $\mathbb{E}|\zeta(0)|^p < \infty$ ($\exists p > 2 \vee 1/(d+1/2)$), fractional parameter $d \in \mathbb{R} \setminus \mathbb{N}_-$ and tempering parameter $\lambda_N > 0$ satisfying (4.2).*

(i) (Strongly tempered errors.) *Let $\lambda_* = \infty$, $d \in \mathbb{R} \setminus \mathbb{N}_-$. Then*

$$\min(1, \lambda_N^{-2d})N(\hat{\beta}_N - 1) \xrightarrow{d} \frac{1}{2 \int_0^1 B^2(s) ds} \begin{cases} B^2(1), & d > 0, \\ B^2(1) - 1, & d = 0, \\ -\Gamma(1-2d)/\Gamma(1-d)^2, & d < 0, \end{cases}$$

where B is a standard Brownian motion.

(ii) (Weakly tempered errors.) *Let $\lambda_* = 0$ and $H = d + \frac{1}{2} \in (0, 1)$. Then*

$$N^{1 \wedge (1+2d)}(\hat{\beta}_N - 1) \xrightarrow{d} \frac{1}{2 \int_0^1 B_H^2(s) ds} \begin{cases} B_H^2(1), & \frac{1}{2} < H < 1, \\ B^2(1) - 1, & H = \frac{1}{2} \\ -H\Gamma(H + \frac{1}{2})/\Gamma(\frac{3}{2} - H), & 0 < H < \frac{1}{2}, \end{cases} \quad (5.10)$$

where B_H is a FBM with variance $\mathbb{E}B_H^2(t) = t^{2H}$, $B = B_{1/2}$.

(iii) (Moderately tempered errors.) *Let $0 < \lambda_* < \infty$ and $H = d + \frac{1}{2} > 0$. Then*

$$N^{1 \wedge (1+2d)}(\hat{\beta}_N - 1) \xrightarrow{d} \frac{1}{2 \int_0^1 (B_{H, \lambda_*}^{II}(s))^2 ds} \begin{cases} (B_{H, \lambda_*}^{II}(1))^2, & H > \frac{1}{2}, \\ (B_{\frac{1}{2}, \lambda_*}^{II}(1))^2 - 1, & H = \frac{1}{2} \\ -\Gamma(2(1-H))/\Gamma(\frac{3}{2} - H)^2, & 0 < H < \frac{1}{2}, \end{cases} \quad (5.11)$$

where $B_{H, \lambda}^{II}$ is a TFBM II given by (3.6).

Remark 5.3 The limit (5.10) in the weakly tempered case coincides with Sowell's limit (5.3). Since $B_{H,0}^{II} \stackrel{\text{fdd}}{=} C_1 B_H$, where $C_1^2 = \Gamma(1-H)/2^{2H} H \Gamma(H+1/2) \sqrt{\pi}$, see [22], the r.v. on the r.h.s. of (5.11) for $\lambda_* = 0, 0 < H < 1$ also coincides with (5.3), however the convergence (5.11) holds for any $H > 0$ in contrast to $H \in (0, 1)$ in (5.10).

Proof of Theorem 5.2. From Theorem 4.3 and Proposition 5.1 we obtain the joint convergence of

$$\left(a_N (S_N^{d, \lambda_N}(1))^2, a_N \int_0^1 (S_N^{d, \lambda_N}(s))^2 ds, b_N \sum_{t=1}^N X_{d, \lambda_N}^2(t) \right) \quad (5.12)$$

where $a_N \rightarrow 0, b_N \rightarrow 0$ are normalizations defined in these theorems and depending on d and λ_* , in each case (i)-(iii) of Theorem 5.2. Then the statement of Theorem 5.2 follows from (5.12), the continuous mapping theorem and the representation of $\hat{\beta}_N - 1$ in (5.4)-(5.6) through the corresponding quantities in (5.12). \square

6 Proofs of Theorem 4.3 and Proposition 5.1

Proof of Theorem 4.3. (i) We restrict the proof of finite-dimensional convergence in (4.6) to one-dimensional convergence at $t > 0$ since the general case follows similarly. We use Proposition 4.2. Accordingly, write $N^{-\frac{1}{\alpha}} \lambda_N^d S_N^{d, \lambda_N}(t) = Q(g_N(t, \cdot)) = \sum_{i \in \mathbb{Z}} g_N(t; i) \zeta(i)$, where $g_N(t; i) := N^{-1/\alpha} \lambda_N^d \sum_{k=1 \vee i}^{[Nt]} e^{-\lambda_N(k-i)} b_d(k-i)$. It suffices to prove (4.5) for suitable p and $g(t; x) := \mathbf{1}_{[0, t]}(x)$. We have

$$\tilde{g}_N(t; x) = \lambda_N^d \sum_{k=1 \vee [Nx]}^{[Nt]} e^{-\lambda_N(k-[Nx])} b_d(k - [Nx]). \quad (6.1)$$

Let us prove the point-wise convergence:

$$\tilde{g}_N(t; x) \rightarrow g(t; x) = \mathbf{1}_{[0, t]}(x), \quad \forall x \neq 0, t. \quad (6.2)$$

Let us prove that conditions (1.2)-(1.4) imply that

$$G_N := \lambda_N^d \sum_{k=0}^{\infty} e^{-\lambda_N k} b_d(k) \rightarrow 1 \quad (N \rightarrow \infty). \quad (6.3)$$

First, let $d > 0$. Then since $\sum_{k=0}^n b_d(k) \sim (1/d\Gamma(d))n^d, n \rightarrow \infty$ according to (1.2), then by applying the Tauberian theorem for power series (Feller [7], Ch. 13, § 5, Thm. 5) we have $\sum_{k=0}^{\infty} b_d(k) e^{-\lambda_N k} \sim (1 - e^{-\lambda_N})^{-d}, N \rightarrow \infty$, proving (6.3) for $d > 0$. Next, let $d = 0$. Then in view of (1.4) the dominated convergence theorem applies yielding $\sum_{k=0}^{\infty} e^{-\lambda_N k} b_0(k) \rightarrow \sum_{k=0}^{\infty} b_0(k) = 1$ and (6.3) follows again.

Next, let $-1 < d < 0$. Then $\tilde{b}_d(k) := \sum_{i=k}^{\infty} b_d(i) \sim (-1/d\Gamma(d))k^{\tilde{d}-1}, k \rightarrow \infty, \tilde{d} := d + 1 \in (0, 1)$, $\tilde{b}_d(0) = 0$ and $\sum_{k=0}^{\infty} e^{-\lambda_N k} b_d(k) = -e^{\lambda_N}(1 - e^{-\lambda_N}) \sum_{k=1}^{\infty} e^{-\lambda_N k} \tilde{b}_d(k)$ using summation by parts. Then the aforementioned Tauberian theorem implies $\sum_{k=0}^{\infty} e^{-\lambda_N k} b_d(k) \sim (1 - e^{-\lambda_N})^{1-\tilde{d}} = (1 - e^{-\lambda_N})^{-d}$ proving (6.3) for $-1 < d < 0$. In the general case $-j < d < -j + 1, j = 1, 2, \dots$ relation (6.3) follows similarly using summation by parts j times.

Let $0 < x < t$ first. Then $\tilde{g}_N(t; x) = G_N - \tilde{g}_N^*(t; x)$, where

$$\tilde{g}_N^*(t; x) := \lambda_N^d \sum_{k > [Nt] - [Nx]} b_d(k) e^{-\lambda_N k}.$$

Using (1.2) for $d \neq 0$ we obtain

$$\begin{aligned} |\tilde{g}_N^*(t; x)| &\leq C(N\lambda_N)^d N^{-1} \sum_{k > [Nt] - [Nx]} e^{-(N\lambda_N)(k/N)} (k/N)^{d-1} \\ &\leq C(N\lambda_N)^d \int_{t-x}^{\infty} e^{-N\lambda_N y} y^{d-1} dy \\ &= C \int_{(t-x)(N\lambda_N)}^{\infty} e^{-z} z^{d-1} dz \rightarrow 0 \end{aligned} \quad (6.4)$$

since $N\lambda_N \rightarrow \infty$. A similar result for $d = 0$ follows directly from (1.4). In view of (6.3), this proves (6.2) for $0 < x < t$. Next, let $x < 0$. Then similarly as above

$$\begin{aligned}\tilde{g}_N(t; x) &\leq C(N\lambda_N)^d N^{-1} \sum_{k > \lfloor Nx \rfloor} e^{-(N\lambda_N)(k/N)} (k/N)^{d-1} \\ &\leq C(N\lambda_N)^d \int_{|x|}^{\infty} e^{-N\lambda_N y} y^{d-1} dy \\ &= C \int_{|x|(N\lambda_N)}^{\infty} e^{-z} z^{d-1} dz \rightarrow 0,\end{aligned}\tag{6.5}$$

proving (6.2). Note also that $|\tilde{g}_N(t; x)| \leq C\lambda_N^d \sum_{k=0}^{\infty} |b_d(k)| \leq C\lambda_N^d \leq C$ for $d < 0$ and $|\tilde{g}_N(t; x)| \leq C(N\lambda_N)^d \int_0^{\infty} e^{-N\lambda_N y} y^{d-1} dy \leq C \int_0^{\infty} e^{-z} z^{d-1} dz \leq C$ for $d > 0$, implying that $|\tilde{g}_N(t; x)|$ is bounded uniformly in $x \in \mathbb{R}, N \geq 1$; moreover, according to (6.5) $|\tilde{g}_N(t; x)| \leq Ce^{-c'|x|}, x < -2$ decays exponentially with $x \rightarrow -\infty$ with some $c' > 0$ uniformly in $N \geq 1$. This proves (4.5) and hence (4.6).

Consider the functional convergence in (4.7). This follows from the tightness criterion

$$N^{-p/2} \lambda_N^{pd} \mathbb{E} |S_N^{d, \lambda_N}(t) - S_N^{d, \lambda_N}(s)|^p \leq C |L_N(t) - L_N(s)|^{p/2}, \quad \forall 0 \leq s < t \leq 1, \tag{6.6}$$

where $L_N(t) := \lfloor Nt \rfloor / N$, see [4], also ([10], Lemma 4.4.1). By Rosenthal's inequality (see e.g. [10], Proposition 4.4.3), $\mathbb{E} |S_N^{d, \lambda_N}(t) - S_N^{d, \lambda_N}(s)|^p \leq C \mathbb{E}^{p/2} |S_N^{d, \lambda_N}(t) - S_N^{d, \lambda_N}(s)|^2$ where $\mathbb{E} |S_N^{d, \lambda_N}(t) - S_N^{d, \lambda_N}(s)|^2 = N\lambda_N^{-2d} \int |\tilde{g}_N(t; x) - \tilde{g}_N(s; x)|^2 dx \leq CN\lambda_N^{-2d} |L_N(t) - L_N(s)|$ follows similarly as above, proving (6.6) and part (i), too.

(ii) Relation (4.8) is well-known with $S_N^{d, \lambda_N}(t)$ replaced by $S_N^{d, 0}(t)$, see e.g. [2], also ([10], Cor. 4.4.1), so that it suffices to prove

$$R_N(t) := S_N^{d, \lambda_N}(t) - S_N^{d, 0}(t) = o_p(N^H). \tag{6.7}$$

With Proposition 4.2 in mind, (6.7) follows from $\|\tilde{g}_N^0(t; \cdot)\|_p \rightarrow 0$, where

$$\begin{aligned}|\tilde{g}_N^0(t; x)| &:= N^{-d} \left| \sum_{k=1 \vee \lfloor Nx \rfloor}^{\lfloor Nt \rfloor} b_d(k - \lfloor Nx \rfloor) (1 - e^{-\lambda_N(k - \lfloor Nx \rfloor)}) \right| \\ &\leq C(N\lambda_N) \frac{1}{N^{d+1}} \sum_{k=1 \vee \lfloor Nx \rfloor}^{\lfloor Nt \rfloor} (k - \lfloor Nx \rfloor)^d \leq C(N\lambda_N) \rightarrow 0\end{aligned}$$

uniformly in $x \in \mathbb{R}$, where we used (1.2), inequality $1 - e^{-x} \leq x$ ($x \geq 0$) and the fact that $H \in (0, 1), 1 < \alpha \leq 2$ imply $-1 < d < 1 - \frac{1}{\alpha}$. For $x < -1$ a similar argument leads to

$$\begin{aligned}|\tilde{g}_N^0(t; x)| &\leq CN^{-d} \sum_{k=1}^{\lfloor Nt \rfloor} (k - \lfloor Nx \rfloor)^{d-1} \leq CN^{-d} ((\lfloor Nt \rfloor - \lfloor Nx \rfloor)^d - (-\lfloor Nx \rfloor)^d) \\ &\leq C((t - x)^d - (-x)^d) \leq C(-x)^{d-1}\end{aligned}\tag{6.8}$$

implying the dominating bound $|\tilde{g}_N^0(t; x)| \leq C/(1+|x|)^{1-d} =: \bar{g}(x)$ where $\|\bar{g}\|_p < \infty$ for $1 \leq p < \alpha$ sufficiently close to α due to condition $d < 1 - \frac{1}{\alpha}$. This proves $\|\tilde{g}_N^0(t; \cdot)\|_p \rightarrow 0$, hence (6.7) and (4.8), too.

To prove the tightness part of (ii), we use a similar criterion as in (6.6), viz.,

$$N^{-pH} \mathbb{E}|S_N^{d, \lambda_N}(t) - S_N^{d, \lambda_N}(s)|^p \leq C|L_N(t) - L_N(s)|^q, \quad \forall 0 \leq s < t \leq 1, \quad (6.9)$$

with $L_N(t) = [Nt]/N$ and suitable $p, q > 1$. Let first $\frac{1}{\alpha} < H < 1$, or $0 < d < 1 - \frac{1}{\alpha}$. Let $\tilde{g}_N(t; x) = N^{-d} \sum_{k=1 \vee [Nx]}^{[Nt]} b_d(k - [Nx]) e^{-\lambda_N(k - [Nx])}$. Then for $0 < s < t$

$$\begin{aligned} |\tilde{g}_N(t; x) - \tilde{g}_N(s; x)| &\leq N^{-d} \sum_{k=[Ns]+1}^{[Nt]} |b_d(k - [Nx])| \leq CN^{-d} \sum_{k=[Ns]+1}^{[Nt]} (k - [Nx])_+^{d-1} \\ &\leq CN^{-d} \int_{[Ns]}^{[Nt]} (y - [Nx])_+^{d-1} dy \\ &\leq C((L_N(t) - L_N(s))_+^d - (L_N(s) - L_N(x))_+^d) \end{aligned} \quad (6.10)$$

and therefore for $1 \leq p < \alpha$ sufficiently close to α

$$\begin{aligned} N^{-pH} \mathbb{E}|S_N^{d, \lambda_N}(t) - S_N^{d, \lambda_N}(s)|^p &\leq C\|\tilde{g}_N(t; \cdot) - \tilde{g}_N(s; \cdot)\|_p^p \\ &\leq C \int_0^\infty |(L_N(t) - L_N(s) + L_N(x))^d - L_N(x)^d|^{p_X} dx \\ &\leq C\left((L_N(t) - L_N(s))^{pd} N^{-1} + \int_0^\infty |(L_N(t) - L_N(s) + x)^d - x^d|^{p_X} dx\right) \\ &\leq C(L_N(t) - L_N(s))^{1+pd} \end{aligned} \quad (6.11)$$

since $L_N(t) - L_N(s) \neq 0$ implies $L_N(t) - L_N(s) \geq N^{-1}$. This proves (6.9) with $q = 1 + pd > 1$.

Next, let $\alpha = 2$ and $0 < H < \frac{1}{2}$, or $-1/2 < d < 0$. Then (6.9) holds for $S_N^{d, 0}$ instead of S_N^{d, λ_N} with $q = pH > 1$, see ([10], proof of Prop. 4.4.4). Hence, it suffices to prove a similar bound for $R_N(t)$ in (6.7). By Rosenthal's inequality (see the proof (6.6)) $\mathbb{E}|R_N(t) - R_N(s)|^p \leq C\mathbb{E}^{p/2}|R_N(t) - R_N(s)|^2$ and hence $N^{-pH} \mathbb{E}|R_N(t) - R_N(s)|^p \leq C\|\tilde{g}_N^0(t; \cdot) - \tilde{g}_N^0(s; \cdot)\|_2^p$, where

$$\begin{aligned} |\tilde{g}_N^0(t; x) - \tilde{g}_N^0(s; x)| &\leq CN^{-d} \sum_{k=[Ns]+1}^{[Nt]} (k - [Nx])_+^{d-1} (1 - e^{-\lambda_N(k - [Nx])}) \\ &\leq C \begin{cases} (-L_N(x))^{d-1} (L_N(t) - L_N(s)), & x < -1, \\ C((L_N(t) - L_N(x))_+^{d+1} - (L_N(s) - L_N(x))_+^{d+1}), & -1 < x < 1, \end{cases} \end{aligned} \quad (6.12)$$

similarly as in (6.10). Hence $N^{-pH} \mathbb{E}|R_N(t) - R_N(s)|^p \leq C\|\tilde{g}_N^0(t; \cdot) - \tilde{g}_N^0(s; \cdot)\|_2^p \leq C(L_N(t) - L_N(s))^{1+p(1+d)}$ follows, proving (6.7) and part (ii), too.

(iii) Similarly as in the proof of (ii), let us prove $\|\tilde{g}_N(t; \cdot) - g(t; \cdot)\|_p \rightarrow 0$, where

$$\tilde{g}_N(t; x) := \frac{1}{N^d} \sum_{k=1 \vee [Nx]}^{[Nt]} b_d(k - [Nx]) e^{-\lambda_N(k - [Nx])}, \quad g(t; x) := \frac{1}{\Gamma(1+d)} h_{H, \alpha, \lambda_*}(t; x), \quad (6.13)$$

see the definition of $h_{H,\alpha,\lambda}$ in (1.9). First, let $d > 0$ or $H > \frac{1}{\alpha}$. Then using (1.2) we obtain the point-wise convergence

$$\begin{aligned}\tilde{g}_N(t; x) &= \frac{1}{N\Gamma(d)} \sum_{\frac{1}{N} \vee \frac{[Nx]}{N} \leq \frac{k}{N} \leq \frac{[Nt]}{N}} \left(\frac{k}{N} - \frac{[Nx]}{N}\right)^{d-1} (1 + \epsilon_{N1}(k, x)) e^{-\lambda_* (\frac{k}{N} - \frac{[Nx]}{N})(1 + \epsilon_{N2})} \\ &\rightarrow \frac{1}{\Gamma(d)} \int_0^t (y - x)_+^{d-1} e^{-\lambda_*(y-x)_+} dy = \frac{1}{\Gamma(1+d)} h_{H,\alpha,\lambda_*}(t; x), \quad \forall x \neq 0, t, \quad (6.14)\end{aligned}$$

see (3.3), where

$$\epsilon_{N1}(k, x) := \Gamma(d)(k - [Nx])^{1-d} b_d(k - [Nx]) - 1 \rightarrow 0, \quad \epsilon_{N2} := (N\lambda_N/\lambda_*) - 1 \rightarrow 0$$

as $N \rightarrow \infty, k - [Nx] \rightarrow \infty$ and $|\epsilon_{N1}(k, x)| + |\epsilon_{N2}| < C$ is bounded uniformly in N, k, x . Therefore (6.14) holds by the dominated convergence theorem. We also have from (6.13) that $|\tilde{g}_N(t; x)| \leq CN^{-d} \sum_{k=1 \vee [Nx]}^{[Nt]} (k - [Nx])^{d-1} e^{-(\lambda_*/2)(k - [Nx])} \leq C \int_0^t (s - x)_+^{d-1} e^{-(\lambda_*/2)(s-x)_+} ds = Ch_{d+1/\alpha, \alpha, \lambda_*/2}(t; x) =: \bar{g}(x)$ is dominated by an integrable function, see (3.3), with $\|\bar{g}\|_p < \infty$. This proves (4.9) for $d > 0$.

Next, let $-\frac{1}{\alpha} < d < 0$. Decompose $\tilde{g}_N(t; x)$ in (6.13) as $\tilde{g}_N(t; x) = \tilde{g}_{N1}(x) - \tilde{g}_{N2}(t; x)$, where

$$\begin{aligned}\tilde{g}_{N1}(x) &:= N^{-d} \sum_{k=1 \vee [Nx]}^{\infty} b_d(k - [Nx]) e^{-\lambda_N(k - [Nx])}, \\ \tilde{g}_{N2}(x) &:= N^{-d} \sum_{[Nt]+1}^{\infty} b_d(k - [Nx]) e^{-\lambda_N(k - [Nx])}.\end{aligned}$$

Let $0 < x < t$. First we have

$$\tilde{g}_{N1}(x) = \frac{\sum_{j=0}^{\infty} b_d(j) e^{-\lambda_N j}}{\lambda_N^{-d}} (N\lambda_N)^{-d} \rightarrow \lambda_*^{-d}$$

since the last ratio tends to 1 as $N \rightarrow \infty$, see (6.3). We also have

$$\begin{aligned}\tilde{g}_{N2}(t; x) &= \frac{1}{N\Gamma(d)} \sum_{\frac{k}{N} > \frac{[Nt]}{N}} \left(\frac{k}{N} - \frac{[Nx]}{N}\right)^{d-1} (1 + \epsilon_{N1}(k, x)) e^{-\lambda_* (\frac{k}{N} - \frac{[Nx]}{N})(1 + \epsilon_{N2})} \\ &\rightarrow \frac{1}{\Gamma(d)} \int_t^{\infty} (s - x)_+^{d-1} e^{-\lambda_*(s-x)_+} ds = \lambda_*^{-d} - \frac{1}{\Gamma(1+d)} h_{H,\alpha,\lambda_*}(t; x),\end{aligned}$$

see (3.4), similarly to (6.14). This proves the point-wise convergence $\tilde{g}_N(t; x) \rightarrow g(t; x) = \Gamma(1+d)^{-1} h_{H,\alpha,\lambda}(t; x)$ for $0 < x < t$ and the proof for $x < 0$ is similar. Then $\|\tilde{g}_N(t; \cdot) - g(t; \cdot)\|_p \rightarrow 0$ or (4.9) for $-\frac{1}{\alpha} < d < 0$ follows similarly as in the case $d > 0$ above.

Consider the proof of tightness in (iii). We use the same criterion (6.9) as in part (ii). Let first $d > 0$. Then for $|x| \leq 1$ the bound in (6.10) and hence $\int_{-1}^1 |\tilde{g}_N(t; x) - \tilde{g}_N(s; x)|^p dx \leq C(L_N(t) - L_N(s))^{1+pd}$ follows as in (6.11). On the other hand, for $x < -1$ we have

$$|\tilde{g}_N(t; x) - \tilde{g}_N(s; x)| \leq C e^{-(\lambda_*/2)|x|} \int_{[Ns/N]}^{[Nt]/N} \frac{1}{y} dy = C e^{-(\lambda_*/2)|x|} (L_N(t) - L_N(s))$$

implying $\int_{-\infty}^{-1} |\tilde{g}_N(t; x) - \tilde{g}_N(s; x)|^p \mathfrak{X} \leq C |L_N(t) - L_N(s)|^p$. Consequently, (6.9) for $d > 0$ holds with $q = (1 + pd) \wedge p > 1$. Finally, (6.9) for $\alpha = 2, -1/2 < d < 0$ follows as in case (ii) since (6.12) holds in the case $\lambda_N = O(1/N)$ as well. This ends the proof of Theorem 4.3. \square

Proof of Proposition 5.1. (i) By stationarity, $N^{-1} \mathbb{E} \sum_{t=1}^N X_{d, \lambda_N}^2(t) = \mathbb{E} X_{d, \lambda_N}^2(0)$. Let us first prove the convergence of expectations:

$$\mathbb{E} X_{d, \lambda_N}^2(0) \rightarrow \frac{\Gamma(1-2d)}{\Gamma^2(1-d)}, \quad d < 1/2, \quad (6.15)$$

$$\lambda_N^{2d-1} \mathbb{E} X_{d, \lambda_N}^2(0) \rightarrow \frac{\Gamma(d-1/2)}{2\sqrt{\pi} \Gamma(d)}, \quad d > 1/2, \quad (6.16)$$

$$|\log \lambda_N|^{-1} \mathbb{E} X_{d, \lambda_N}^2(0) \rightarrow \frac{1}{\pi}, \quad d = 1/2, \quad (6.17)$$

as $N \rightarrow \infty$. Since $\mathbb{E} X_{d, \lambda_N}^2(0) = \sum_{k=0}^{\infty} e^{-2\lambda_N k} \omega_{-d}^2(k)$, with $\omega_{-d}(k) \equiv \omega_{-d}(k)$ defined in (2.3), and

$$\sum_{k=0}^n \omega_{-d}^2(k) \sim \begin{cases} (1/(2d-1)\Gamma^2(d))n^{2d-1}, & d > 1/2, \\ (1/\Gamma^2(1/2))\log(n), & d = 1/2, \\ \sum_{k=0}^{\infty} \omega_{-d}^2(k) = \Gamma(1-2d)/\Gamma^2(1-d), & d < 1/2 \end{cases}$$

as $n \rightarrow \infty$, the convergences in (6.15)-(6.17) follows from the Tauberian theorem in [7] used in the proof of Theorem 4.3 (i) above.

With (6.15)-(6.17) in mind, (5.7)-(5.9) follow from

$$Q_N \equiv N^{-1} \sum_{t=1}^N \left[X_{d, \lambda_N}^2(t) - \mathbb{E} X_{d, \lambda_N}^2(t) \right] = \begin{cases} o_p(1), & d < 1/2, \\ o_p(\lambda_N^{1-2d}), & d > 1/2, \\ o_p(|\log \lambda_N|), & d = 1/2. \end{cases} \quad (6.18)$$

Let $\omega_{-d, \lambda}(k) := \omega_{-d}(k)e^{-\lambda k}$. We have $Q_N = Q_{N1} + Q_{N2}$, where

$$\begin{aligned} Q_{N1} &= N^{-1} \sum_{s \leq N} (\zeta^2(s) - \mathbb{E} \zeta^2(s)) \sum_{t=1 \vee s}^N \omega_{-d, \lambda_N}^2(t-s), \\ Q_{N2} &= N^{-1} \sum_{s_2 < s_1 \leq N} \zeta(s_1) \sum_{t=1 \vee s_1}^N \omega_{-d, \lambda_N}(t-s_1) \omega_{-d, \lambda_N}(t-s_2) \zeta(s_2). \end{aligned}$$

Note $Q_{Ni}, i = 1, 2$ are sums of martingale differences. We shall use the well-known moment inequality for sums of martingale differences:

$$\mathbb{E} \left| \sum_{i \geq 1} \xi_i \right|^\alpha \leq 2 \sum_{i \geq 1} \mathbb{E} |\xi_i|^\alpha \quad (6.19)$$

see e.g. ([10], Prop. 2.5.2), which is valid for any $1 \leq \alpha \leq 2$ and any sequence $\{\xi_i, i \geq 1\}$ with $\mathbb{E} |\xi_i|^\alpha < \infty, \mathbb{E} [\xi_i | \xi_j, 1 \leq j < i] = 0, i \geq 1$.

First, let $d > 1/2$. Using $\mathbb{E}|\zeta(0)|^p < \infty$, $2 < p < 4$ and (6.19) with $\alpha = p/2$ we obtain

$$\begin{aligned}
\mathbb{E}|Q_{N1}|^{p/2} &\leq CN^{-p/2} \sum_{s \leq N} \left| \sum_{t=1 \vee s}^N \omega_{-d, \lambda_N}^2(t-s) \right|^{p/2} \\
&\leq CN^{-p/2} \sum_{s \leq N} \left(\sum_{t=1 \vee s}^N (t-s)_+^{2(d-1)} e^{-2\lambda_N(t-s)} \right)^{p/2} \\
&\leq CN^{-p/2} \left\{ \int_N^\infty \mathfrak{s} \left(\int_0^N (t+s)^{2(d-1)} e^{-2\lambda_N(t+s)} \mathfrak{t} \right)^{p/2} + N \left(\int_0^N t^{2(d-1)} e^{-2\lambda_N t} \mathfrak{t} \right)^{p/2} \right\} \\
&\equiv CN^{-p/2} \{I_{N1} + I_{N2}\}.
\end{aligned}$$

Here, $I_{N2} \leq N \left(\int_0^\infty t^{2(d-1)} e^{-2\lambda_N t} \mathfrak{t} \right)^{p/2} = CN \lambda_N^{(p/2)(1-2d)}$ and $I_{N1} \leq CN^{p/2} \int_N^\infty s^{(d-1)p} e^{-\lambda_N p s} \mathfrak{s} \leq CN^{p/2} \lambda_N^{(1-d)p-1}$. Therefore,

$$\mathbb{E}|Q_{N1}|^{p/2} \leq C(\lambda_N^{(1-d)p-1} + N^{1-p/2} \lambda_N^{(p/2)(1-2d)}). \quad (6.20)$$

Next,

$$\begin{aligned}
\mathbb{E}|Q_{N2}|^2 &\leq CN^{-2} \sum_{s_2 < s_1 \leq N} \mathbb{E} \left(\sum_{t=1 \vee s_1}^N \omega_{-d, \lambda_N}(t-s_1) \omega_{-d, \lambda_N}(t-s_2) \zeta(s_2) \right)^2 \\
&\leq CN^{-2} \sum_{s_2 < s_1 \leq N} \sum_{t=1 \vee s_1}^N \omega_{-d, \lambda_N}^2(t-s_1) \omega_{-d, \lambda_N}^2(t-s_2) \\
&\leq CN^{-2} \sum_{t=1}^N \left(\sum_{s \leq t} \omega_{-d, \lambda_N}^2(t-s) \right)^2 \\
&\leq CN^{-2} \int_0^N \mathfrak{t} \left(\int_{-\infty}^t (t-s)^{2(d-1)} e^{-2\lambda_N(t-s)} \mathfrak{s} \right)^2 \\
&= CN^{-1} \left(\int_0^\infty s^{2(d-1)} e^{-2\lambda_N s} \mathfrak{s} \right)^2 \leq CN^{-1} \lambda_N^{2(1-2d)}. \quad (6.21)
\end{aligned}$$

Since $(1-d)p-1 > (p/2)(1-2d)$ and $1-(p/2) < 0$, (6.20) and (6.21) prove (6.18) for $d > 1/2$.

Next, let $d < 1/2$. Then similarly as above we obtain

$$\begin{aligned}
\mathbb{E}|Q_{N1}|^{p/2} &\leq CN^{-p/2} \sum_{s \leq N} \left(\sum_{t=1 \vee s}^\infty (t-s)_+^{2(d-1)} \right)^{p/2} \\
&= CN^{-p/2} \left\{ N \left(\sum_{t=1}^N t^{2(d-1)} \right)^{p/2} + \sum_{s \geq N} \left(\sum_{t=1}^N (s+t)^{2(d-1)} \right)^{p/2} \right\} \\
&\leq CN^{1-p/2} + CN^{-p/2} \sum_{s \geq N} (Ns^{2(d-1)})^{p/2} \leq CN^{1-p/2} \quad (6.22)
\end{aligned}$$

and

$$\begin{aligned}
\mathbb{E}|Q_{N2}|^2 &\leq CN^{-2} \sum_{s_2 < s_1 \leq N} \sum_{t=1 \vee s_1}^N \omega_{-d, \lambda_N}^2(t-s_1) \omega_{-d, \lambda_N}^2(t-s_2) \\
&\leq CN^{-2} \sum_{t=1}^N \left(\sum_{s < t} (t-s)^{2(d-1)} \right)^2 \leq CN^{-1}. \quad (6.23)
\end{aligned}$$

(6.22) and (6.23) prove (6.18) for $d < 1/2$.

Finally, let $d = 1/2$. Then since $p > 2$

$$\begin{aligned}
\mathbb{E}|Q_{N1}|^{p/2} &\leq CN^{-p/2} \sum_{s \leq N} \left(\sum_{t=1 \vee s}^{\infty} (t-s)_+^{-1} \right)^{p/2} \\
&= CN^{-p/2} \left\{ N \left(\sum_{t=1}^N t^{-1} \right)^{p/2} + \sum_{s \geq N} \left(\sum_{t=1}^N (s+t)^{-1} \right)^{p/2} \right\} \\
&\leq CN^{1-p/2} (\log N)^{p/2} + CN^{-p/2} \sum_{s \geq N} (Ns^{-1})^{p/2} \\
&\leq CN^{1-p/2} (\log N)^{p/2} = o(1)
\end{aligned} \tag{6.24}$$

while

$$\mathbb{E}|Q_{N2}|^2 \leq CN^{-2} \sum_{t=1}^N \left(\sum_{s < t} (t-s)^{-1} e^{-\lambda_N(t-s)} \right)^2 \leq CN^{-1} (\log \lambda_N)^2. \tag{6.25}$$

(6.24) and (6.25) prove (6.18) for $d = 1/2$. Proposition 5.1 is proved. \square

References

- [1] Abramowitz, M., Stegun, I.: Handbook of mathematical functions, ninth edition, Dover, New York (1965).
- [2] Astrauskas, A. (1983). Limit theorems for sums of linearly generated random variables. *Lithuanian J. Math.* **23** 127–134.
- [3] Balan, R., Jakubowski, A. and Louhichi, S. (2016). Functional convergence of linear processes with heavy-tailed innovations. *J. Theoret. Probab.* **29** 491–526.
- [4] Billingsley, P. (1968). *Convergence of Probability Measures*. New York: Wiley.
- [5] Brockwell, P.J. and Davis, R.A. (1991). *Time Series: Theory and Methods*, 2nd ed.. New York: Springer.
- [6] Dickey, D. and Fuller, W. (1979). Distribution of the estimators for autoregressive time series with a unit root. *JASA* **74** 427–431.
- [7] Feller, W. (1966). *An Introduction to Probability Theory and Its Applications, vol. 2*. New York: Wiley.
- [8] Giraitis, L., Kokoszka, P. and Leipus, R. (2000). Stationary ARCH models: dependence structure and central limit theorem. *Econometric Theory* **16** 3–22.
- [9] Giraitis, L., Kokoszka, P., Leipus, R. and Teyssière, G. (2003). Rescaled variance and related tests for long memory in volatility and levels. *J. Econometrics* **112** 265–294.
- [10] Giraitis, L., Koul, H.L. and Surgailis, D. (2012). *Large Sample Inference for Long Memory Processes*. London: Imperial College Press.
- [11] Gradshteyn, I.S. and Ryzhik, I.M. (2000). *Tables of Integrals and Products*. 6th edition. New York: Academic Press.
- [12] Granger, C.W.J. and Joyeux, R. (1980). An introduction to long-memory time series models and fractional differencing. *J. Time Series Anal.* **1** 15–29.

- [13] Hosking, J.R.M. (1981). Fractional differencing. *Biometrika* **68** 165–176.
- [14] Kasahara, Y. and Maejima, M. (1988). Weighted sums of i.i.d. random variables attracted to integrals of stable processes. *Probab. Theory Relat. Fields* **78** 75–96.
- [15] Kokoszka, P.S. and Taqqu, M.S. (1995). Fractional ARIMA with stable innovations. *Stochastic Process. Appl.* **60** 19–47.
- [16] Lavancier, F., Leipus, R., Philippe, A. and Surgailis, D. (2013). Detection of non-constant long memory parameter. *Econometric Theory* **29** 1009–1056.
- [17] Lo, A. (1991). Long-term memory in stock market prices. *Econometrica* **59** 1279–1313.
- [18] Meerschaert, M.M. and Sabzikar, F. (2013). Tempered fractional Brownian motion. *Statist. Probab. Lett.* **83** 2269–2275.
- [19] Meerschaert, M.M. and Sabzikar, F. (2016). Tempered fractional stable motion. *J. Theoret. Probab.* **29** 681–706.
- [20] Meerschaert, M.M., Sabzikar, F., Phanikumar, M.S. and Zeleke, A. (2014). Tempered fractional time series model for turbulence in geophysical flows. *J. Stat. Mech. Theory Exp.* **2014** P09023.
- [21] Phillips, P.C.B. (1987). Time series regression with a unit root. *Econometrica* **55** 277–301.
- [22] Sabzikar, F. and Surgailis, D. (2017). Tempered fractional Brownian and stable motions of second kind. Preprint. Available on <http://arxiv.org/abs/1702.07258>
- [23] Samorodnitsky, G. and Taqqu, M.S. (1996). *Stable Non-Gaussian Random Processes: Stochastic Models with Infinite Variance*. Boca Raton etc: Chapman and Hall.
- [24] Sowell, F. (1990). The fractional unit root distribution. *Econometrica* **58** 495–505.